

# 具有时滞的递归神经网络模型的分支分析

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**摘要:**主要研究了一类含有6个时滞的四阶神经网络模型的分支问题. 通过应用时滞微分方程的中心流形定理和规范型理论, 得到了系统在原点处的 Bogdanov-Takens (B-T) 分支和 triple zero 分支的规范型, 进而给出了一些主要的分支现象.

**关键词:**神经网络模型; B-T 分支; triple zero 分支; 规范型

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## 1 分支的存在性

近几年, 各个领域的学者和科研工作者都对神经网络模型十分感兴趣<sup>[1-5]</sup>, 在文献[5]中主要研究了含3个神经元神经网络模型的 Hopf 分支, 共振和非共振的双 Hopf 分支, 受该文的启发, 本文主要考虑如下的含有4个神经元神经网络模型的 B-T 分支和 triple zero 分支问题.

$$\begin{cases} \dot{x}_1 = -x_1(t) + f(x_2(t-\tau_2)), \dot{x}_2 = -x_2(t) + f(x_3(t-\tau_4)), \dot{x}_3 = -x_3(t) + f(x_4(t-\tau_6)), \\ \dot{x}_4 = -x_4(t) + af(x_1(t-\tau_1)) + bf(x_2(t-\tau_3)) + cf(x_3(t-\tau_5)), \end{cases} \quad (1)$$

其中  $x_i (i = 1, 2, 3, 4)$  表示第  $i$  个神经元的状态,  $a, b, c$  表示连接权,  $\tau_i$  是非负时滞,  $f(x)$  是神经元间诱发的非线性函数, 且满足  $f(0) = 0, f'(0) = 1$ .

令  $u_1(t) = x_1(t), u_2(t) = x_2(t-\tau_2), u_3(t) = x_3(t-\tau_2-\tau_4), u_4(t) = x_4(t-\tau_2-\tau_4-\tau_6)$ , 则系统(1)可以写成如下形式

$$\begin{cases} \dot{u}_1 = -u_1(t) + f(u_2(t)), \dot{u}_2 = -u_2(t) + f(u_3(t)), \dot{u}_3 = -u_3(t) + f(u_4(t)), \\ \dot{u}_4 = -u_4(t) + af(u_1(t-\sigma_1)) + bf(u_2(t-\sigma_2)) + cf(u_3(t-\sigma_3)), \end{cases} \quad (2)$$

其中  $\sigma_1 = \tau_1 + \tau_2 + \tau_4 + \tau_6, \sigma_2 = \tau_3 + \tau_4 + \tau_6, \sigma_3 = \tau_5 + \tau_6$ .

不失一般性, 假设  $\tau_1 > \tau_2 > \tau_3 > \tau_4 > \tau_5 > \tau_6 > 0$ , 可得  $\sigma_1 > \sigma_2 > \sigma_3$ . 系统(2)在原点处的特征方程是

$$F(\lambda) = (\lambda + 1)^4 - ae^{-\lambda\sigma_1} - (\lambda + 1)be^{-\lambda\sigma_2} - (\lambda + 1)^2ce^{-\lambda\sigma_3} = 0. \quad (3)$$

由方程(3)可以得到

$$\begin{aligned} F(0) &= 1 - a - b - c, F'(0) = 4 - b - 2c + a\sigma_1 + b\sigma_2 + c\sigma_3, \\ F''(0) &= 12 - 2c - a\sigma_1^2 + 2b\sigma_2 - b\sigma_2^2 + 4c\sigma_3 - c\sigma_3^2. \end{aligned} \quad (4)$$

从  $F(0) = F'(0) = 0$  解得

$$a = a_0 = \frac{(c-1)\sigma_2 - \sigma_3 + c - 3}{\sigma_1 - \sigma_2 + 1}, b = b_0 = -\frac{(c-1)\sigma_1 - \sigma_3 + 2c - 4}{\sigma_1 - \sigma_2 + 1}. \quad (5)$$

通过  $F''(0) = 0$ , 得出

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$$c = c_0 = \frac{(\sigma_2 + 3)\sigma_1^2 + (12 - \sigma_2^2 + 2\sigma_2)\sigma_1 + 12 - 4\sigma_2^2 - 4\sigma_2}{(\sigma_2 - \sigma_3 + 1)\sigma_1^2 + m_0\sigma_1 + (\sigma_3 - 2)\sigma_1^2 + (2 - \sigma_3^2 + 2\sigma_3)\sigma_2 + \sigma_3^2 - 4\sigma_3 + 2},$$

其中  $m_0 = -\sigma_2^2 + 2\sigma_2 + \sigma_3^2 - 4\sigma_3 + 2$ .

因此,可以计算出  $F'''(0) \neq 24 + a\sigma_1^3 - 3b\sigma_2^2 + b\sigma_3^2 + 6c\sigma_3 - 6c\sigma_2^2 + c\sigma_3^3$ ,假设(H0) $F'''(0) \neq 0$ 成立.因此,可总结如下引理.

**引理 1** i) 如果(H1) $a = a_0, b = b_0, c \neq c_0$ 成立,则  $\lambda = 0$ 是方程(3)的二重零根;(ii) 如果(H0)和(H2) $a = a_0, b = b_0, c = c_0$ 成立,则  $\lambda = 0$ 是方程(3)的三重零根.

## 2 B-T 分支的规范型

为了讨论系统(2)在原点处的B-T分支,把  $a_0$ 和  $b_0$ 当作分支参数.因此,令  $a = a_0 + \mu_1, b = b_0 + \mu_2$ ,得到下面的系统:

$$\begin{cases} \dot{u}_1 = -u_1(t) + f(u_2(t)), \dot{u}_2 = -u_2(t) + f(u_3(t)), \dot{u}_3 = -u_3(t) + f(u_4(t)), \\ \dot{u}_4 = -u_4(t) + (a_0 + \mu_1)f(u_1(t - \sigma_1)) + (b_0 + \mu_2)f(u_2(t - \sigma_2)) + cf(u_3(t - \sigma_3)), \end{cases} \quad (6)$$

其中  $\mu = (\mu_1, \mu_2)$ 在原点(0,0)附近充分小.令  $k = f''(0)$ ,系统(6)的泰勒展开式如下:

$$\dot{U}(t) = AU(t) + \sum_{i=1}^3 B_i U(t - \sigma_i) + \frac{1}{2} \hat{F}_2(U, P) + \dots, \quad (7)$$

为了获得系统(6)在原点处的规范型,需要计算系统(7)的广义特征空间  $P$ 的基  $\Phi(\theta)$ 及其对偶空间  $P^*$ 的基  $\Psi(s)$ ,其中  $\Phi(\theta)$ 和  $\Psi(s)$ 的选取详见文献[6].通过计算可得

$$\Phi(\theta) = \begin{pmatrix} 1 & \theta \\ 1 & \theta + 1 \\ 1 & \theta + 2 \\ 1 & \theta + 3 \end{pmatrix}, \Psi(0) = \begin{pmatrix} v_{11} & v_{12} & v_{13} & v_{14} \\ v_{21} & v_{22} & v_{23} & v_{24} \end{pmatrix}, \quad (8)$$

且  $\Phi = \Psi J_0, \Psi = -J_0 \Psi$ 和  $\langle \Phi, \Psi \rangle = I$ 成立,其中  $J_0 = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ ,

$$\begin{aligned} v_{11} &= \frac{2n_0(\alpha_2 - \alpha_3 + c - \sigma_2 - 3)}{3l^2}, v_{13} = \frac{(\sigma_1 - \sigma_2 + 1)[v_{11} + (\sigma_1 + 2)v_{21}]}{\alpha_2 - \alpha_3 + c - \sigma_2 - 3}, \\ v_{12} &= \frac{(\sigma_1 - \sigma_2 + 1)[(c - 1)v_{11} + (\alpha_1 - \alpha_3 + c - \sigma_1 - 3)v_{21}]}{-\alpha_2 + \alpha_3 - c + \sigma_2 + 3}, \\ v_{14} &= \frac{(-\sigma_1 + \sigma_2 - 1)[v_{11} + (\sigma_1 + 1)v_{21}]}{-\alpha_2 + \alpha_3 - c + \sigma_2 + 3}, v_{21} = -\frac{2(\alpha_2 - \alpha_3 + c - \sigma_2 - 3)}{l}, \\ v_{22} &= \frac{2(\sigma_1 - \sigma_2 + 1)(c - 1)}{l}, v_{23} = v_{24} = -\frac{2(\sigma_1 - \sigma_2 + 1)}{l}, \end{aligned}$$

$$\begin{aligned} l &= (\alpha_2 - \alpha_3 + c - \sigma_2 - 3)\sigma_1^2 + [(1 - c)\sigma_2^2 + (2c - 2)\sigma_2 + \alpha_3^2 - 4\alpha_3 + 2c - 12]\sigma_1 + \\ &\quad (\alpha_3 - 2c + 4)\sigma_2^2 + (-\alpha_3^2 + 2\alpha_3 + 2c + 4)\sigma_2 + \alpha_3^2 - 4\alpha_3 + 2c - 12, \\ n_0 &= \{(2\sigma_2 - 2\sigma_3 + 2)\sigma_1^3 + (-3\sigma_2^2 + 3\sigma_3^2 + 9\sigma_2 - 15\sigma_3 + 9)\sigma_1^2 + [\sigma_2^3 + (3\sigma_3 - 12)\sigma_2^2 + \\ &\quad (-3\sigma_3^2 + 6\sigma_3 + 12)\sigma_2 - \sigma_3^3 + \sigma_3^2 - \sigma_3 + 12]\sigma_1 + (2 - \sigma_3)\sigma_2^3 + 6(\sigma_3 - 2)\sigma_2^2 + \\ &\quad (\sigma_3^3 - 9\sigma_3^2 + 12\sigma_3 + 6)\sigma_2 - \sigma_3^3 + 9\sigma_3^2 - 18\sigma_3 + 6\}c - 2(\sigma_2 + 3)\sigma_1^3 + \\ &\quad (3\sigma_2^2 - 9\sigma_2 - 45)\sigma_1^2 + (-\sigma_2^2 + 18\sigma_2^2 + 6\sigma_2 - 96)\sigma_1 - 4\sigma_2^3 + 24\sigma_2^2 + 36\sigma_2 - 60. \end{aligned}$$

令  $z = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$ .根据文献[7-8],系统(6)可被分解成

$$\begin{aligned} \dot{z} &= J_0 z + \frac{1}{2!} f_2^1(z, y, \mu) + \frac{1}{3!} f_3^1(z, y, \mu) + \dots, \\ \dot{y} &= A_0 y + \frac{1}{2!} f_2^2(z, y, \mu) + \frac{1}{3!} f_3^2(z, y, \mu) + \dots. \end{aligned} \quad (9)$$

其中  $f_j^1 = \Psi(0)F(\Phi x + y, \mu), f_j^2 = (I - \pi)X_0 F(\Phi x + y, \mu), j = 1, 2$ .然后,根据文献[7],通过应用空间

$\text{Im}M_2^c$  和  $(\text{Im}M_2^c)^\perp$  的基,定义  $P_{I,2^p} = a$  满足  $p - a \in (\text{Im}M_2^c)^\perp$ . 对任意的  $p \in V_2^1(\mathbb{R}_2)$ , 知  $\text{Proj}(\text{Im}M_2^c)^\perp(p)$  有以下几种情况:

$$\text{Proj}_{(\text{Im}M_2^c)^\perp}(p) = \begin{cases} p, p \in \text{Im}(M_2^c), \\ 0, p \in \text{Im}(M_2^c), \end{cases} \quad \text{Proj}_{(\text{Im}M_2^c)^\perp} \begin{pmatrix} 0 \\ 2z_1 z_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 2z_1 z_2 \end{pmatrix},$$

$$\text{Proj}_{(\text{Im}M_2^c)^\perp} \begin{pmatrix} z_1 \mu_1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ z_2 \mu_1 \end{pmatrix}, \text{Proj}_{(\text{Im}M_2^c)^\perp} \begin{pmatrix} z_1 \mu_2 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ z_2 \mu_2 \end{pmatrix}.$$

因此,系统在  $B-T$  奇点处的规范型有如下的形式:

$$\dot{z} = J_0 z + \frac{1}{2} g_2^1(z, 0, \mu) + \frac{1}{3} g_3^1(z, 0, \mu) + \dots, \quad (10)$$

其中  $g_j^1(z, 0, \mu)$  是在  $(z, \mu)$  上的  $j(j \geq 2)$  次齐次多项式,且  $g_j^1 = (I - P_I, j^1) \bar{f}_j^1 (j \geq 2)$  参考文献([7,9]). 由  $\varphi = \Phi(\theta)z$  和(8)可得出

$$\begin{aligned} \varphi_1(0) &= z_1 + z_2, \varphi_1(-\sigma_1) = z_1 + (1 - \sigma_1)z_2, \varphi_2(0) = z_1 + 2z_2, \\ \varphi_2(-\sigma_2) &= z_1 + (2 - \sigma_2)z_2, \varphi_3(0) = z_1 + 3z_2, \varphi_3(-\sigma_3) = z_1 + (3 - \sigma_3)z_2, \varphi_4(0) = z_1 + 4z_2. \end{aligned} \quad (11)$$

然后,由(8),(9)和(11),得到

$$\begin{aligned} \frac{1}{2} f_2^1(\tilde{x}, 0, \mu) &= \Psi(0) \left[ \frac{1}{2} \tilde{F}_2(\tilde{\Phi}\tilde{x}, \mu) \right] = \begin{pmatrix} v_{11} & v_{12} & v_{13} & v_{14} \\ v_{21} & v_{22} & v_{23} & v_{24} \end{pmatrix} \times \\ &\begin{pmatrix} \frac{k}{2} \varphi_2^2(0) \\ \frac{k}{2} \varphi_3^2(0) \\ \frac{k}{2} \varphi_4^2(0) \\ \sum_{i=1}^2 \mu_i \varphi_i(-\sigma_i) + \frac{a_0 k}{2} \varphi_1^2(-\sigma_1) + \frac{b_0 k}{2} \varphi_2^2(-\sigma_2) + \frac{ck}{2} \varphi_3^2(-\sigma_3) \\ \begin{pmatrix} a_{2000} z_1^2 + a_{1100} z_1 z_2 + v_{14} z_1 \mu_1 + v_{14} z_1 \mu_2 + a_{0200} z_2^2 + \\ v_{14} (\sigma_1 + 1) z_2 \mu_1 + v_{14} (\sigma_2 + 2) z_2 \mu_2 + b_{2000} z_1^2 + \\ b_{1100} z_1 z_2 + v_{24} z_1 \mu_1 + v_{24} z_1 \mu_2 + b_{0200} z_2^2 + \\ v_{24} (\sigma_1 + 1) z_2 \mu_1 + v_{24} (\sigma_2 + 2) z_2 \mu_2 \end{pmatrix} \end{pmatrix}, \end{aligned} \quad (12)$$

其中

$$\begin{aligned} a_{2000} &= \frac{k}{2} (v_{11} + v_{12} + v_{13} + v_{14}), a_{0200} = \frac{k}{2} (4v_{11} + 9v_{12} + 16v_{13}) + \frac{ke_1 v_{14}}{2(\sigma_1 + 1 - \sigma_2)}, \\ a_{1100} &= (2v_{11} + 3v_{12} + 4v_{13})k + \frac{k(2\sigma_1 - 4\sigma_2 + 2\sigma_3 - \sigma_1 + 3\sigma_2 + 5)v_{14}}{\sigma_1 + 1 - \sigma_2}, \\ b_{2000} &= \frac{k}{2} (v_{21} + v_{22} + v_{23} + v_{24}), b_{0200} = \frac{k}{2} (4v_{21} + 9v_{22} + 16v_{23}) + \frac{ke_1 v_{24}}{2(\sigma_1 + 1 - \sigma_2)}, \\ b_{1100} &= (2v_{21} + 3v_{22} + 4v_{23})k + \frac{k(2\sigma_1 - 4\sigma_2 + 2\sigma_3 - \sigma_1 + 3\sigma_2 + 5)v_{24}}{\sigma_1 + 1 - \sigma_2}, \\ e_1 &= [(\sigma_2 - \sigma_3)(\sigma_1 - \sigma_3)(\sigma_1 - \sigma_2) + \sigma_1^2 - 2\sigma_1\sigma_2 + 4\sigma_3\sigma_1 - 2\sigma_2^2 - 2\sigma_2\sigma_3 + \sigma_3^2 + 7\sigma_1 - \\ &16\sigma_2 + 9\sigma_3 + 2]c - \sigma_1^2\sigma_2 + \sigma_1\sigma_2^2 - 3\sigma_1^2 + 2\sigma_1\sigma_2 + 4\sigma_2^2 - 2\sigma_1 + 15\sigma_2 + 13. \end{aligned}$$

因此,从系统(10),可以得到

$$\dot{z}_1 = z_2, \dot{z}_2 = \lambda_1 z_1 + \lambda_2 z_2 + \gamma_1 z_1^2 + \gamma_2 z_1 z_2 + \dots, \quad (13)$$

其中

$$\begin{aligned} \lambda_1 &= v_{24}(\mu_1 + \mu_2), \lambda_2 = [v_{14} + v_{24}(\sigma_1 + 1)]\mu_1 + [v_{14} + v_{24}(\sigma_2 + 2)]\mu_2, \gamma_1 = b_{2000}, \\ \gamma_2 &= 2a_{2000} + b_{1100} = \frac{2k[sol - (\sigma_1 - 2\sigma_2 + \sigma_3 - 3\sigma_1 + 4\sigma_2)n_0]}{3l^2}, \end{aligned}$$

$$s_0 = 3(\sigma_1 + 1 - \sigma_2)(\sigma_1 - \sigma_3 + 2c - 3\sigma_1 - 12).$$

进一步, 易得到  $\left| \frac{\partial(\lambda_1, \lambda_2)}{\partial(\mu_1, \mu_2)} \right| = v_{24}^2(\sigma_1 - \sigma_2 + 1) > 0$ .

因此, 映射  $(\mu_1, \mu_2) \rightarrow (\lambda_1, \lambda_2)$  是正则的. 可以得出如下的定理.

**定理 1** 如果  $k = f''(0) \neq 0$ , 系统(7) 在中心流形上等价于系统(13).

参考文献[6], (13) 式有两个平衡点  $E_1 = (0, 0)$  和  $E_2 = (-\frac{\lambda_1}{\gamma_1}, 0)$ , 在 origin 附近以  $\mu_1$  和  $\mu_2$  为分支参数的分支曲线分析如下:

- 1) (13) 式在曲线  $S = \{(\lambda_1, \lambda_2) : \lambda_1 = 0, \lambda_2 \in \mathbb{R}\} = \{(\mu_1, \mu_2) : v_2 4(\mu_1 + \mu_2) = 0\}$  经历一个超临界分支;
- 2) 在平凡平衡点  $E_1$  处, (13) 式在半曲线  $H = \{(\lambda_1, \lambda_2) : \lambda_2 = 0, \lambda_1 < 0\} = \{(\mu_1, \mu_2) : [v_1 4 + v_{24}(\sigma_1 + 1)]\mu_1 + [v_{14} + v_{24}(\sigma_2 + 2)]\mu_2 = 0, v_{24}(\mu_1 + \mu_2) < 0\}$  上经历了一个不稳定的 Hopf 分支;
- 3) 在平衡点  $E_2$  处, (13) 式在半曲线  $T = \{(\lambda_1, \lambda_2) : \lambda_2 = \frac{\gamma_2}{\gamma_1}\lambda_1, \lambda_1 > 0\} = \{(\mu_1, \mu_2) : [v_{14} + v_{24}(\sigma_1 + 1)]\mu_1 + [v_{14} + v_{24}(\sigma_2 + 2)]\mu_2 = \frac{\gamma_2}{\gamma_1}v_{24}(\mu_1 + \mu_2), v_{24}(\mu_1 + \mu_2) > 0\}$  上经历了一个稳定的 Hopf 分支.

### 3 Triple zero 分支的规范型

在这一部分, 把  $a_0, b_0, c_0$  看作分支参数来讨论系统(2) 在 origin 处的三重分支. 令  $a = a_0 + \beta_1, = b_0 + \beta_2, c = c_0 + \beta_3$ , 由此, 得到如下的系统:

$$\dot{u}_1 = -u_1(t) + f(u_2(t)), \dot{u}_2 = -u_2(t) + f(u_3(t)), \dot{u}_3 = -u_3(t) + f(u_4(t)),$$

$$\dot{u}_4 = -u_4(t) + (a_0 + \beta_1)f(u_1(t - \sigma_1)) + (b_0 + \beta_2)f(u_2(t - \sigma_2)) + (c_0 + \beta_3)f(u_3(t - \sigma_3)), \quad (14)$$

其中  $\beta = (\beta_1, \beta_2, \beta_3)$  在 origin  $(0, 0, 0)$  附近充分小. 同样, 系统(14) 的泰勒展开式如下:

$$\dot{U}(t) = AU(t) + \sum_{i=1}^2 B_i U(t - \sigma_i) + \tilde{B}_3 U(t - \sigma_3) + \frac{1}{2} \tilde{F}_2(U, P) + \dots \quad (15)$$

为了获得系统(6) 在 origin 处的规范型, 需要计算系统(7) 的广义特征空间  $P$  的基  $\Phi(\theta)$  及其对偶空间  $P^*$  的基  $\Psi(s)$ . 其中  $\Phi(\theta)$  和  $\Psi(s)$  的选取详见文献[10] 的引理 3.1. 通过计算, 可以得出

$$\Phi(\theta) = \begin{pmatrix} 1 & \theta & \frac{1}{2}\theta^2 + u_{31} \\ 1 & \theta + 1 & \frac{1}{2}\theta^2 + \theta + u_{31} \\ 1 & \theta + 2 & \frac{1}{2}\theta^2 + 2\theta + u_{31} + 1 \\ 1 & \theta + 3 & \frac{1}{2}\theta^2 + 3\theta + u_{31} + 3 \end{pmatrix}, \Psi(0) = \begin{pmatrix} 0 & v_{12} & v_{13} & v_{14} \\ v_{21} & v_{22} & v_{23} & v_{24} \\ v_{31} & v_{32} & v_{33} & v_{34} \end{pmatrix}, \quad (16)$$

其中  $\Phi(\theta)$  满足  $\dot{\Phi} = \Phi B$ , 且  $B = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ ,

$$u_{31} = -\frac{s_3 v_{21} + s_4 v_{31}}{20s_5 v_{31}}, v_{31} = \frac{6a_0}{a_0 \sigma_1^3 + b_0 \sigma_2^3 + c_0 \sigma_3^3 - 3b_0 \sigma_2^2 - 6c_0 \sigma_3^2 + 6c_0 \sigma_3 + 24},$$

$$v_{32} = \frac{(a_0 + b_0)v_{31}}{a_0}, v_{33} = v_{34} = \frac{v_{31}}{a_0}, v_{21} = -\frac{3a_0 s_2}{2(a_0 \sigma_1^3 + b_0 \sigma_2^3 + c_0 \sigma_3^3 - 3b_0 \sigma_2^2 - 6c_0 \sigma_3^2 + 6c_0 \sigma_3 + 24t)^2},$$

$$v_{22} = \frac{(a_0 + b_0)v_{21} - (a_0 - b_0\sigma_1 + b_0\sigma_2)v_{31}}{a_0}, v_{23} = \frac{(2 + \sigma_1)v_{31} + v_{21}}{a_0},$$

$$v_{24} = \frac{(1 + \sigma_1)v_{31} + v_{21}}{a_0}, v_{12} = -\frac{[a_0 + b_0(\sigma_2 - \sigma_1)]v_{21}}{a_0} + \frac{[b_0(\sigma_1 - \sigma_2)^2 + 2a_0]v_{31}}{2a_0},$$

$$v_{13} = \frac{2(\sigma_1 + 2)v_{21} + (\sigma_1^2 + 4\sigma_1 + 2)v_{31}}{2a_0}, v_{14} = \frac{2(1 + \sigma_1)v_{21} + \sigma_1(\sigma_1 + 2)v_{31}}{2a_0},$$

$$\begin{aligned}
 s_2 &= 3a_0\sigma_1^4 + 4b_0\sigma_1\sigma_2^3 - b_0\sigma_2^4 + 4c_0\sigma_1\sigma_3^3 - c_0\sigma_3^4 + 4a_0\sigma_1^3 - 12b_0\sigma_1\sigma_2^2 + 8b_0\sigma_2^3 - 24c_0\sigma_1\sigma_3^2 + \\
 &\quad 12c_0\sigma_3^3 - 12b_0\sigma_2^2 + 24c_0\sigma_1\sigma_3 - 36c_0\sigma_3^2 + 24c_0\sigma_3 + 96\sigma_1 + 120, \\
 s_3 &= 15a_0\sigma_1^4 + 20b_0\sigma_1\sigma_2^3 - 5b_0\sigma_2^4 + 20c_0\sigma_1\sigma_3^3 - 5c_0\sigma_3^4 + 20a_0\sigma_1^3 - 60b_0\sigma_1\sigma_2^2 + 40b_0\sigma_2^3 - \\
 &\quad 120c_0\sigma_1\sigma_3^2 + 60c_0\sigma_3^3 - 60b_0\sigma_2^2 + 120c_0\sigma_1\sigma_3 - 180c_0\sigma_3^2 + 120c_0\sigma_3 + 480\sigma_1 + 600, \\
 s_4 &= 6a_0\sigma_1^5 + 10b_0\sigma_1^2\sigma_2^3 - 5b_0\sigma_1\sigma_2^4 + b_0\sigma_2^5 + 10c_0\sigma_1^2\sigma_3^3 - 5c_0\sigma_1\sigma_3^4 + c_0\sigma_3^5 + 15a_0\sigma_1^4 - \\
 &\quad 30b_0\sigma_1^2\sigma_2^2 + 40b_0\sigma_1\sigma_2^3 - 10b_0\sigma_2^4 - 60c_0\sigma_1^2\sigma_3^2 + 60c_0\sigma_1\sigma_3^3 - 15c_0\sigma_3^4 - 60b_0\sigma_1\sigma_2^2 + \\
 &\quad 20b_0\sigma_2^3 + 60c_0\sigma_1^2\sigma_3 - 180c_0\sigma_1\sigma_3^2 + 60c_0\sigma_3^3 + 120c_0\sigma_1\sigma_3 - 60c_0\sigma_3^2 + 240\sigma_1^2 + 600\sigma_1 + 120, \\
 s_5 &= a_0\sigma_1^3 + 3b_0\sigma_1^2\sigma_2 - 3b_0\sigma_1\sigma_2^2 + b_0\sigma_2^3 + 3c_0\sigma_1^2\sigma_3 - 3c_0\sigma_1\sigma_3^2 + c_0\sigma_3^3 + 3a_0\sigma_1^2 + 3b_0\sigma_1^2 + \\
 &\quad 6c_0\sigma_1\sigma_3 - 3c_0\sigma_3^2 + 6\sigma_1^2 + 6a_0 + 18\sigma_1 + 6.
 \end{aligned}$$

假设  $\text{Re}\lambda \neq 0$ , 则系统(15) 能被写成

$$\dot{z} = Bz + \frac{1}{2}g_2^1(z, 0, \beta) + \dots, \tag{17}$$

其中  $g_2^1(z, 0, \beta) = (I - P_1, 2^1)f_2^1(\tilde{x}, 0, \mu)$ ,  $f_2^1(\tilde{x}, 0, \mu) = \Psi(0)\tilde{F}(\Phi x, \beta)$ . 因此, 由(15), (16) 和(17) 得到

$$\frac{1}{2}f_2^1(\tilde{x}, 0, \mu) = \psi(0) \left[ \frac{1}{2}\tilde{F}_2(\Phi \hat{x}, \mu) \right] = \begin{pmatrix} 0 & v_{12} & v_{13} & v_{14} \\ v_{21} & v_{22} & v_{23} & v_{24} \\ v_{31} & v_{32} & v_{33} & v_{34} \end{pmatrix} \begin{pmatrix} \frac{k}{2}\varphi_2^2(0) \\ \frac{k}{3}\varphi_2^2(0) \\ \frac{k}{4}\varphi_2^2(0) \\ \beta_1\varphi_1(-\sigma_1) + \beta_2\varphi_2(-\sigma_2) + \beta_3\varphi_3(-\sigma_3) + \\ \frac{a_0k}{2}\varphi_1^2(-\sigma_1) + \frac{b_0k}{2}\varphi_2^2(-\sigma_2) + \frac{c_0k}{2}\varphi_3^2(-\sigma_3) \end{pmatrix} = \begin{pmatrix} a_{200}x_1^2 + a_{110}x_1x_2 + a_{101}x_1x_3 + v_{14}(\beta_2 + \beta_1 + \beta_3)x_1 + a_{020}x_2^2 + \\ a_{011}x_2x_3 + v_{14}[\beta_2(1 - \sigma_2) + \beta_3(2 - \sigma_3) - \beta_1\sigma_1]x_2 + a_{002}x_3^2 + \\ v_{14}[\beta_1(\frac{1}{2}\sigma_1^2 + u_{31}) + \beta_2(\frac{1}{2}\sigma_2^2 - \sigma_2 + u_{31}) + p_0\beta_3]x_3 \\ b_{200}x_1^2 + b_{110}x_1x_2 + b_{101}x_1x_3 + v_{24}(\beta_2 + \beta_1 + \beta_3)x_1 + b_{020}x_2^2 + \\ b_{011}x_2x_3 + v_{24}[\beta_2(1 - \sigma_2) + \beta_3(2 - \sigma_3) - \beta_1\sigma_1]x_2 + b_{002}x_3^2 + \\ v_{24}[\beta_1(\frac{1}{2}\sigma_1^2 + u_{31}) + \beta_2(\frac{1}{2}\sigma_2^2 - \sigma_2 + u_{31}) + p_0\beta_3]x_3 \\ c_{200}x_1^2 + c_{110}x_1x_2 + c_{101}x_1x_3 + v_{34}(\beta_2 + \beta_1 + \beta_3)x_1 + c_{020}x_2^2 + \\ c_{011}x_2x_3 + v_{34}[\beta_2(1 - \sigma_2) + \beta_3(2 - \sigma_3) - \beta_1\sigma_1]x_2 + c_{002}x_3^2 + \\ v_{34}[\beta_1(\frac{1}{2}\sigma_1^2 + u_{31}) + \beta_2(\frac{1}{2}\sigma_2^2 - \sigma_2 + u_{31}) + p_0\beta_3]x_3 \end{pmatrix}, \tag{18}$$

其中

$$\begin{aligned}
 a_{200} &= \frac{k}{2}(v_{12} + v_{13} + v_{14}), a_{110} = k(2v_{12} + 3v_{13} + 4v_{14}), \\
 a_{101} &= k[(u_{31} + 1)v_{12} + (u_{31} + 3)v_{13} + v_{14}(u_{31} + 6)], \\
 a_{020} &= \frac{k}{2}[(\sigma_2^2 - 2\sigma_2 + 1)b_0 + (\sigma_3^2 - 4\sigma_3 + 4)c_0 + a_0\sigma_1^2]v_{14} + 4v_{12} + 9v_{13}, \\
 a_{011} &= \frac{k}{2}[4(u_{31} + 1)v_{12} + 6(u_{31} + 3)v_{13} - (\sigma_1(\sigma_1^2 + 2u_{31})a_0 + \\
 &\quad (\sigma_2 - 1)(\sigma_2^2 + 2u_{31} - 2\sigma_2)b_0 + (\sigma_3 - 2)(\sigma_3^2 + 2u_{31} - 4\sigma_3 + 2)c_0)v_{14}], \\
 a_{002} &= \frac{k}{8}[4(u_{31} + 1)^2v_{12} + 4(u_{31} + 3)^2v_{13} + ((\sigma_1^2 + 2u_{31})^2a_0 + \\
 &\quad (\sigma_2^2 + 2u_{31} - 2\sigma_2)^2b_0 + (\sigma_3^2 + 2u_{31} - 4\sigma_3 + 2)^2c_0)v_{14}],
 \end{aligned}$$

$$\begin{aligned}
b_{200} &= \frac{k}{2}(v_{21} + v_{22} + v_{23} + v_{24}), b_{110} = k(v_{21} + 2v_{22} + 3v_{23} + 4v_{24}), \\
b_{101} &= k[u_{31}v_{21} + (u_{31} + 1)v_{22} + (u_{31} + 3)v_{23} + (u_{31} + 6t)v_{24}], \\
b_{020} &= \frac{k}{2}[(\sigma_2^2 - 2\sigma_2 + 1)b_0 + (\sigma_3^2 - 4\sigma_3 + 4)c_0 + a_0\sigma_1^2]v_{24} + 9v_{23} + 4v_{22} + v_{21}], \\
b_{011} &= \frac{k}{2}[2u_{31}v_{21} + 4(u_{31} + 1)v_{22} + 6(u_{31} + 3)v_{23} - (\sigma_1(\sigma_1^2 + 2u_{31})a_0 + \\
&(\sigma_2 - 1)(\sigma_2^2 + 2u_{31} - 2\sigma_2)b_0 + (\sigma_3 - 2)(\sigma_3^2 + 2u_{31} - 4\sigma_3 + 2)c_0]v_{24}], \\
b_{002} &= \frac{k}{8}[4(1 + u_{31})^2v_{22} + 4(3 + u_{31})^2v_{23} + ((\sigma_1^2 + 2u_{31})^2a_0 + (\sigma_2^2 + \\
&2u_{31} - 2\sigma_2)^2b_0 + c_0(\sigma_3^2 + 2u_{31} - 4\sigma_3 + 2)^2)v_{24} + 4u_{31}^2v_{21}], \\
c_{200} &= \frac{k}{2}(v_{31} + v_{32} + v_{33} + v_{34}), c_{110} = k(v_{31} + 2v_{32} + 3v_{33} + 4v_{34}), \\
c_{101} &= k[u_{31}v_{31} + (1 + u_{31})v_{32} + (3 + u_{31})v_{33} + (u_{31} + 6)v_{34}], \\
c_{020} &= \frac{k}{2}[(\sigma_2^2 - 2\sigma_2 + 1)b_0 + (\sigma_3^2 - 4\sigma_3 + 4)c_0 + a_0\sigma_1^2]v_{34} + 9v_{33} + 4v_{32} + v_{31}], \\
c_{002} &= \frac{k}{8}[4(1 + u_{31})^2v_{32} + 4(3 + u_{31})^2v_{33} + ((\sigma_1^2 + 2u_{31})^2a_0 + (\sigma_2^2 + 2u_{31} - 2\sigma_2)^2b_0 + \\
&c_0(\sigma_3^2 + 2u_{31} - 4\sigma_3 + 2)^2)v_{34} + 4u_{31}^2v_{31}], p_0 = \frac{1}{2}\sigma_3^2 - 2\sigma_3 + u_{31} + 1.
\end{aligned}$$

由系统(17)和(18),可以得到

$$\dot{x}_1 = x_2, \dot{x}_2 = x_3, \dot{x}_3 = k_1x_1 + k_2x_2 + k_3x_3 + h_1x_1^2 + h_2x_2^2 + h_3x_1x_2 + h_4x_1x_3 + \dots \quad (19)$$

其中

$$\begin{aligned}
k_1 &= v_{34}(\beta_1 + \beta_2 + \beta_3), k_2 = (v_{24} - v_{34}\sigma_1)\beta_1 + (v_{34}(1 - \sigma_2) + v_{24})\beta_2 + (v_{34}(2 - \sigma_3) + v_{24})\beta_3, \\
k_3 &= [v_{34}(\frac{1}{2}\sigma_1^2 + u_{31}) + v_{14} - v_{24}\sigma_1]\beta_1 + [(\frac{1}{2}\sigma_2^2 - \sigma_2 + u_{31})v_{34} + (1 - \sigma_2)v_{24} + v_{14}]\beta_2 + \\
&[v_{34}(\frac{1}{2}\sigma_3^2 - 2\sigma_3 + u_{31} + 1) + v_{14} + (2 - \sigma_3)v_{24}]\beta_3, h_1 = c_{200}, h_2 = 2a_{200} + b_{110} + c_{020}, \\
h_3 &= 2b_{200} + c_{110}, h_4 = a_{101} + 2a_{200} + b_{110}.
\end{aligned}$$

容易计算出

$$\left| \frac{\partial(k_1, k_2, k_3)}{\partial(\beta_1, \beta_2, \beta_3)} \right|_{\beta=0} = \frac{108[(\sigma_2 - \sigma_3 + 1)\sigma_1^2 + g_1\sigma_1 + g_0]^4}{(d_1\sigma_1^3 + d_2\sigma_1^2 + d_3\sigma_1 + d_4)^3}, \quad (20)$$

其中

$$\begin{aligned}
g_0 &= (\sigma_3 - 2)\sigma_2^2 + (-\sigma_3^2 + 2\sigma_3 + 2)\sigma_2 + \sigma_3^2 - 4\sigma_3 + 2, g_1 = -\sigma_2^2 + \sigma_3^2 + 2\sigma_2 - 4\sigma_3 + 2, \\
d_1 &= (\sigma_3 + 2)\sigma_2^2 + (-\sigma_3^2 + 2\sigma_3 + 6)\sigma_2 - 3\sigma_3^2 + 6, \\
d_2 &= -(\sigma_3 + 2)\sigma_2^2 + 3(\sigma_3 + 2)\sigma_2^2 + (\sigma_3^3 - 6\sigma_3^2 + 6\sigma_3 + 24)\sigma_2 + 3\sigma_3^3 - 18\sigma_3^2 - 6\sigma_3 + 24, \\
d_3 &= (\sigma_3^2 - 4\sigma_3 - 10)\sigma_2^2 + (-\sigma_3^3 + 3\sigma_3^2 + 6\sigma_3 + 6)\sigma_2^2 + 2(\sigma_3^3 - 6\sigma_3^2 + 6\sigma_3 + 24)\sigma_2 + \\
&12\sigma_3^3 - 48\sigma_3^2 - 24\sigma_3 + 48, \\
d_4 &= 4(\sigma_3^2 - \sigma_3 - 4)\sigma_2^2 + 4\sigma_3(-\sigma_3^2 + 3\sigma_3 + 3)\sigma_2^2 + 4(-\sigma_3^3 + 6\sigma_3 + 12)\sigma_2 + 12\sigma_3^3 - 48\sigma_3^2 - 24\sigma_3 + 48.
\end{aligned}$$

令  $G(\sigma_1) = (\sigma_2 - \sigma_3 + 1)\sigma_1^2 + g_1\sigma_1 + g_0$ , 则  $G(\sigma_1) = 0$  的判别式为

$$\Delta = g_1^2 - 4(\sigma_2 - \sigma_3 + 1)g_0 = [(\sigma_2 - \sigma_3)^2 - 2][(\sigma_2 - \sigma_3 + 4)(\sigma_2 - \sigma_3) + 2],$$

因为  $\sigma_2 > \sigma_3 > 0$ , 可以推出  $\sigma_2 - \sigma_3 + 1 > 0, (\sigma_2 - \sigma_3 + 4)(\sigma_2 - \sigma_3) > 0$ . 其次, 假设  $0 < \sigma_2 - \sigma_3 < \sqrt{2}$ , 得到  $(\sigma_2 - \sigma_3)^2 - 2 < 0$ . 因此,  $\Delta < 0$ , 由此知  $G$  没有实根, 即  $\left| \frac{\partial(k_1, k_2, k_3)}{\partial(\beta_1, \beta_2, \beta_3)} \right|_{\beta=0} \neq 0$ .

因此,可以得到下面的定理.

**定理 2** 如果  $k = f''(0) \neq 0$ , 且引理 1 成立, 则系统(19)在中心流形上等价于系统(14).

根据文献[11], 系统(19)在原点处的分支分析如下:

- 1) 当  $T = \{(\beta_1, \beta_2, \beta_3): k_1 = 0\}$  时, 系统(19) 经历一个超临界分支.
  - 2) 当  $H_1 = \{(\beta_1, \beta_2, \beta_3): k_3 = -\frac{k_1}{k_2}, k_2 < 0\}$  时, 系统(19) 经历一个 Hopf 分支.
  - 3) 当  $H_2 = \left\{(\beta_1, \beta_2, \beta_3): k_3 = \left(\frac{h_4}{h_1} - \frac{h_1}{h_3 k_1 - h_1 k_2}\right) k_1, \frac{h_1}{h_3 k_1 - h_1 k_2} > 0\right\}$  时, 系统(19) 在平衡点  $(-\frac{g_1}{h_1}, 0, 0)$  处经历一个 Hopf 分支.
  - (4) 当  $BT = \{(\beta_1, \beta_2, \beta_3): k_1 = 0, k_2 = 0\}$  时, 系统(19) 经历一个 B-T 分支.
  - (5) 当  $H_3 = \{(\beta_1, \beta_2, \beta_3): k_1 = 0, k_3 = 0, k_2 < 0\}$  时, 系统(19) 经历一个 zero-Hopf 分支.
- 如果  $k = f''(0) = 0$ , 则  $\gamma_1 = \gamma_2 = 0, h_1 = h_2 = 0$ , 即系统(13), (19) 是退化的. 此时, 为了分析系统(2) 在 B-T 奇点和 triplezero 奇点附近的动力学性质, 需要进一步计算相应的三阶规范型. 该问题将在以后的文章中做进一步探讨.

### 参 考 文 献

- [1] Ge J H, Xu J. Stability switches and fold-Hopf bifurcations in an inertial four-neuron network model with coupling delay[J]. Neurocomputing, 2013, 110: 70-79.
- [2] Liu X. Zero singularity of codimension two or three in a four-neuron BAM neural network model with multiple delays[J]. Nonlinear Dyn, 2014, 77: 1783-1794.
- [3] Zhao H Y, Yuan J L, Zhang X B. Stability and bifurcation analysis of reaction-diffusion neural network with delays[J]. Neurocomputing, 2015, 147: 280-290.
- [4] Abdjelil A, Jiang H J. The existence and stability of the anti-periodic solution for delayed Cohen-Grossberg neural network with impulsive effects[J]. Neurocomputing, 2015, 149: 22-28.
- [5] Ding Y T, Jiang W H, Yu P. Bifurcation analysis in a recurrent neural network model with delays[J]. Commun Nonlinear Sci Numer Simulat, 2013, 18: 351-372.
- [6] Dong T, Liao X F. Bogdanov-Takens bifurcation in a tri-neuron BAM neural network model with multiple delays[J]. Nonlinear Dyn, 2013, 71: 583-595.
- [7] Xu Y X, Huang M Y. Homoclinic orbits and Hopf bifurcations in delay differential systems with T-B singularity[J]. J Differential Equations, 2008, 244: 582-598.
- [8] Zhang C H, Yan X P. Bogdanov-Takens singularity in a network of two coupled neurons with delayed feedbacks[J]. J Appl Math Comput, 2009, 29: 469-479.
- [9] Faria T, Magalhães LT. Normal form for retarded functional differential equations and application to Bogdanov-Takens singularity[J]. J Differential Equations, 1995, 122: 201-224.
- [10] Qiao Z Q, Liu X B, Zhu D M. Bifurcation in delay differential systems with triple-zero singularity[J]. Chinese Journal of Contemporary Mathematics, 2010, 31: 59-70.
- [11] Campbell S A, Yuan Y. Zero singularities of codimension two and three in delay differential equations[J]. Nonlinearity, 2008, 21: 2671-2691.

## Bifurcation Analysis in a Recurrent Neural Network Model with Delays

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**Abstract:** In this paper, we pay our main attention to study the bifurcation analysis of a four-node recurrent neural network model with six discrete delays. By using center manifold reduction and normal form method of delay differential equations, the norm forms of the Bogdanov-Takens (B-T) and triple zero bifurcations at origin are obtained. Finally, we gave some main bifurcation phenomena.

**Keywords:** neural network model; B-T bifurcation; triple zero bifurcation; normal form